Disturbance and Regrowth In Biophysical Terms

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Outline

 Complementary set of analyses in biophysical space

Why is biomass the key variable?

 What is our approach to modeling biomass and biomass change?

Why Biomass?

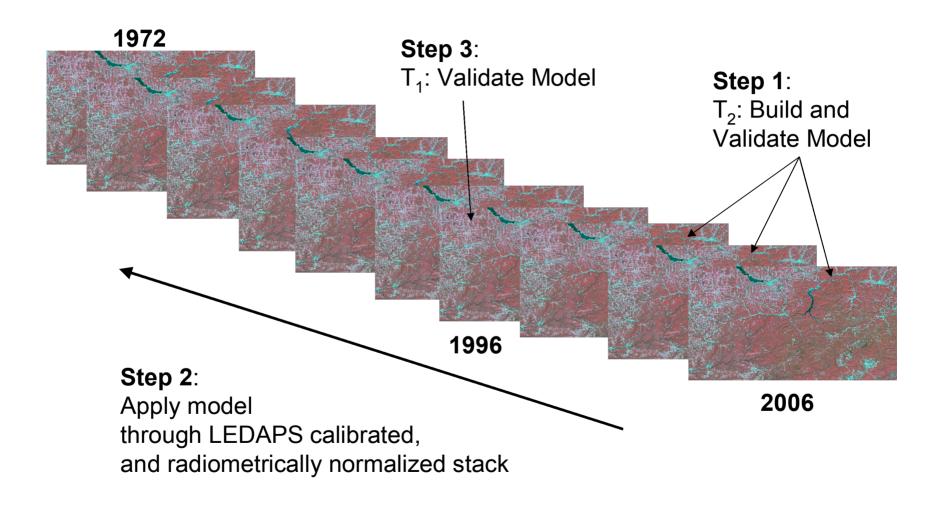
- Link data cubes with biophysical measure
- Directly related to carbon stocks and flux
 - Importance of spatially explicit biomass maps
 - Biomass lost from disturbance and rate of regrowth after disturbance
- Parameterization and validation for process-based models
- Leverages FIA data

Empirical Biomass Modeling

 Limitations of modeling biomass with single date of Landsat imagery are well known

 We are leveraging temporal density to observe biomass trends and minimize error

Biomass Modeling Approach



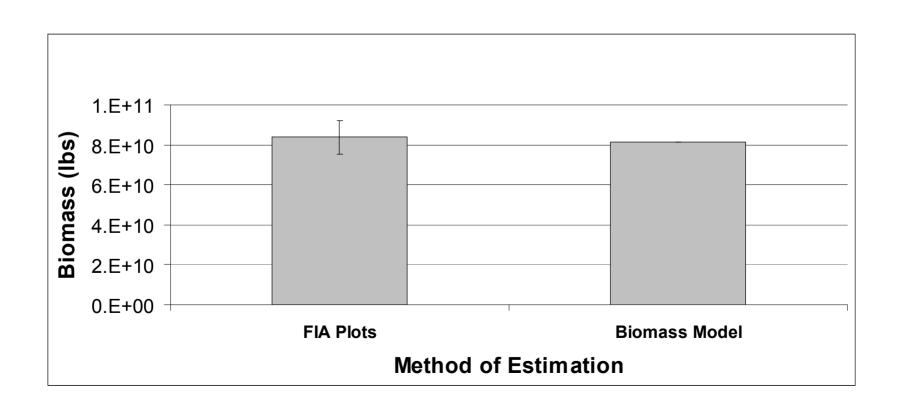
Biomass Modeling Comparison

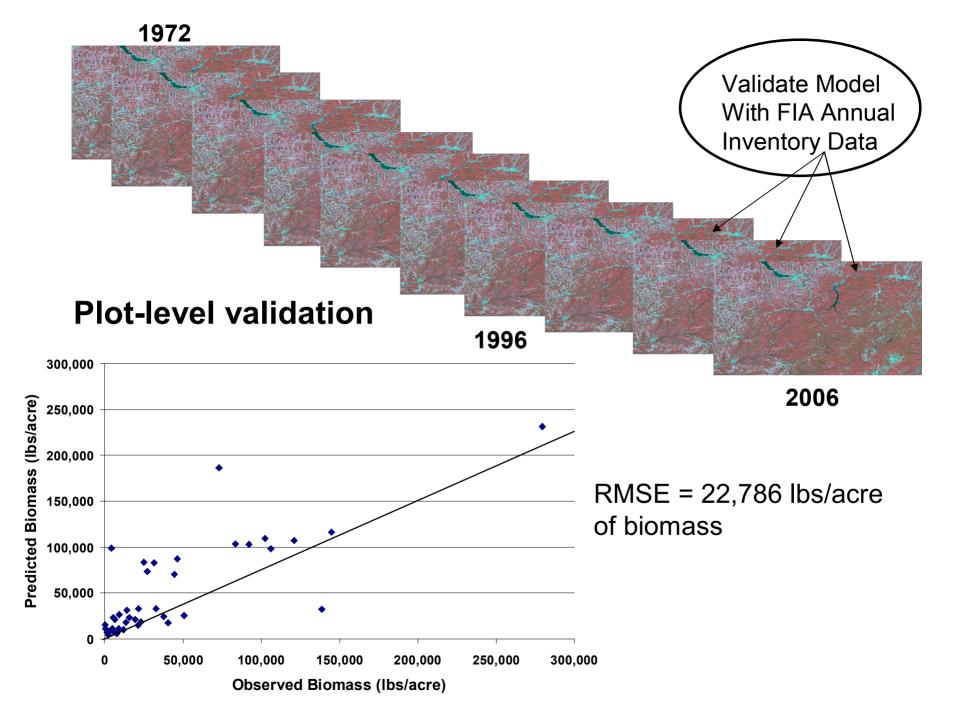
- We compared statistical modeling techniques:
 - Gradient Nearest Neighbors (GNN)
 - Reduced Major Axis Regression (RMA)
 - Generalized Additive Models (GAMs)
 - Stochastic Gradient Boosting (SGB)
 - Regression Tree Analysis (Cubist)

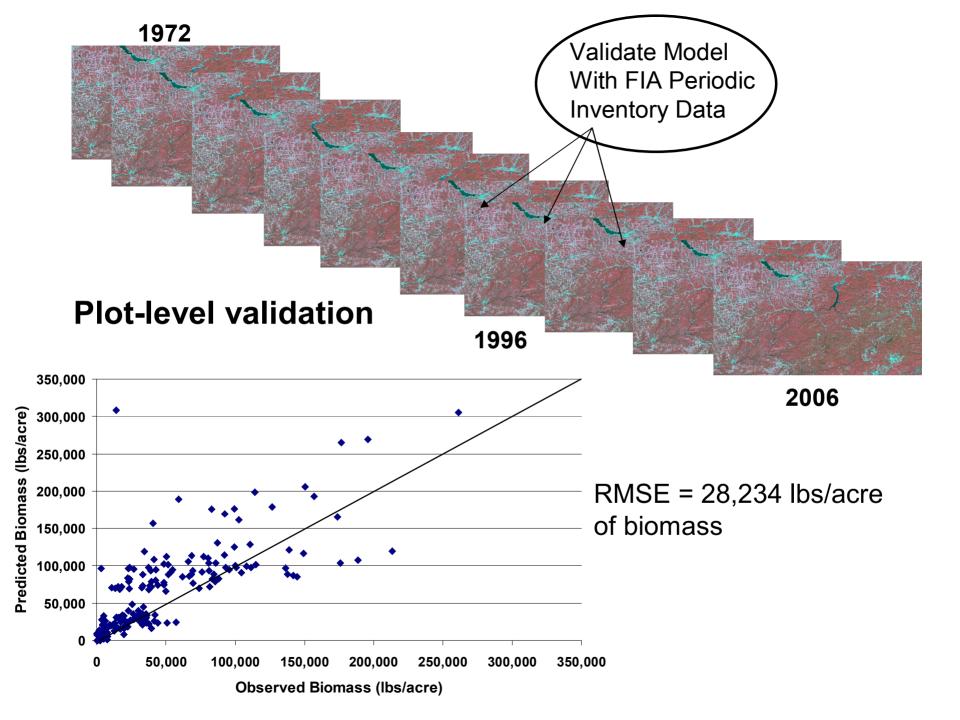
Biomass Modeling Comparison

- Regression Tree Analysis
 - Most accurate models in terms of RMSE
 - Non-parametric
 - Logical interpretation
 - Easy implementation across large number of data cubes

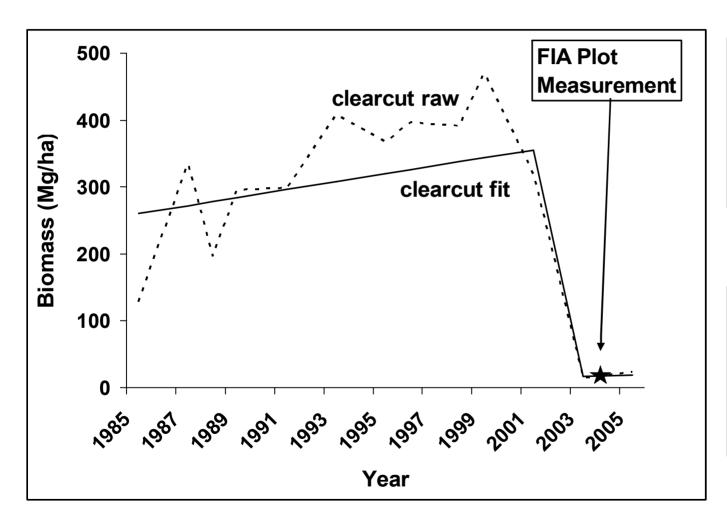
Scene-Level Validation Total Aboveground Forest Biomass Arizona 37/35

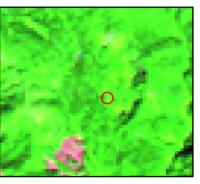






Biomass Trajectory: Clearcut



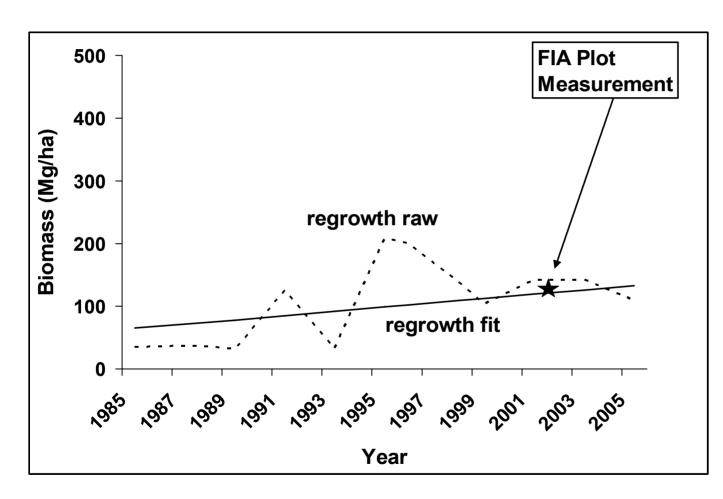


Pre-Clearcut



Post-Clearcut

Biomass Trajectory: Regrowth



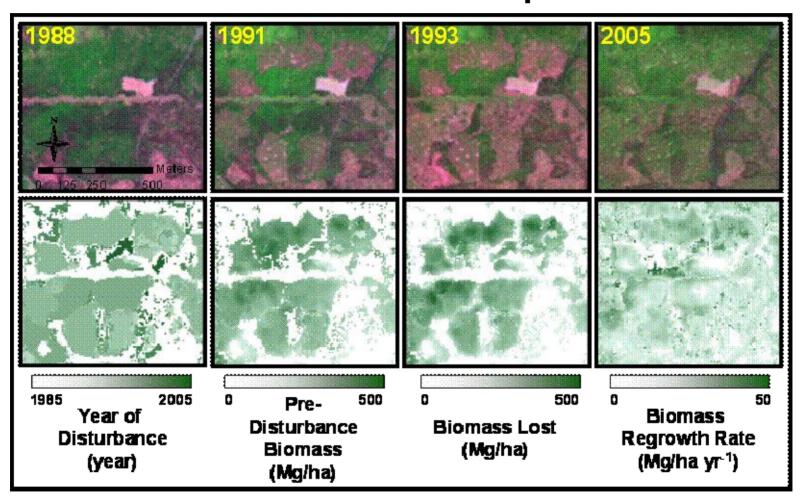


Regrowth in 1985



Regrowth in 2005

Biomass Trajectory-Based Change Detection: Outputs



Summary

 Modeling biomass through time increases the carbon relevance of the disturbance and regrowth analysis

 This is made possible by linking FIA data, the dense Landsat time series, and trajectory-based change detection